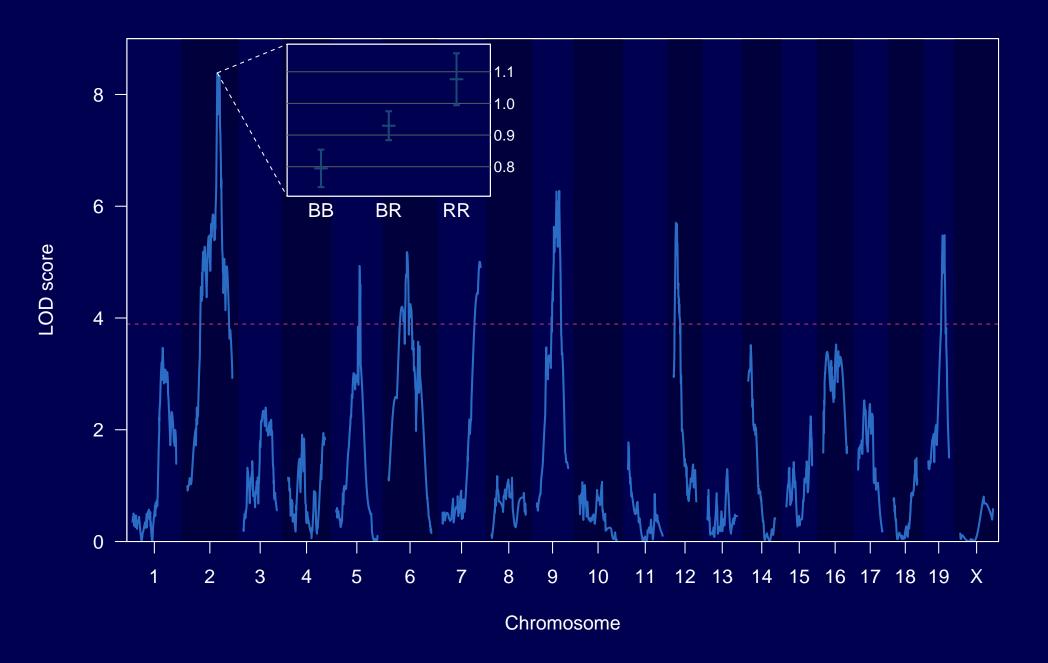
Multiparent populations & R/qtl2

Karl Broman

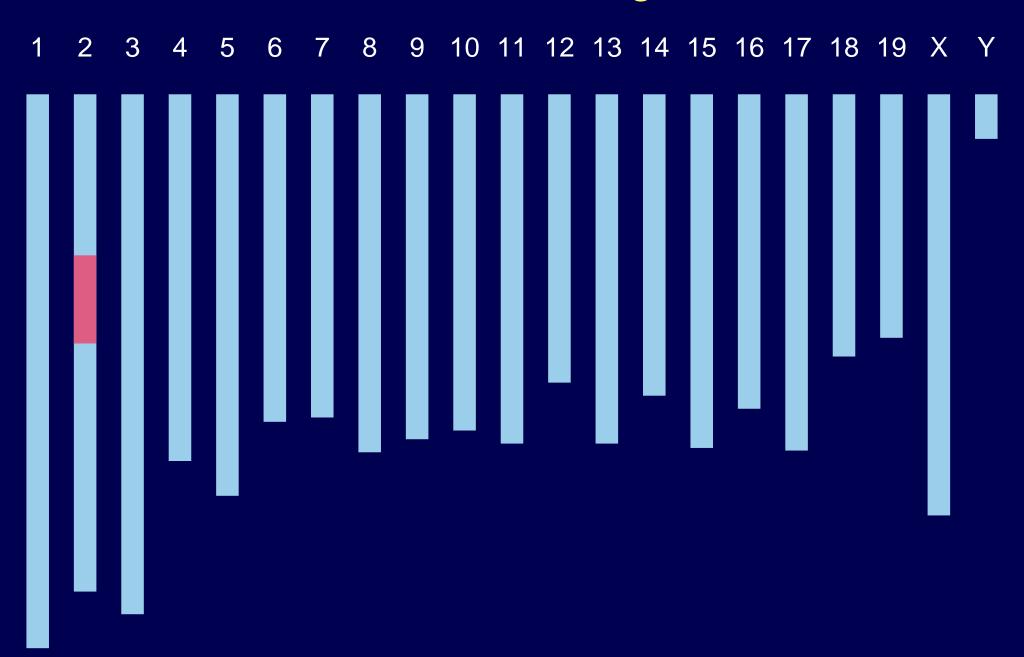
Biostatistics and Medical Informatics University of Wisconsin – Madison

> kbroman.org/qt12 kbroman.org github.com/kbroman @kwbroman

QTL mapping



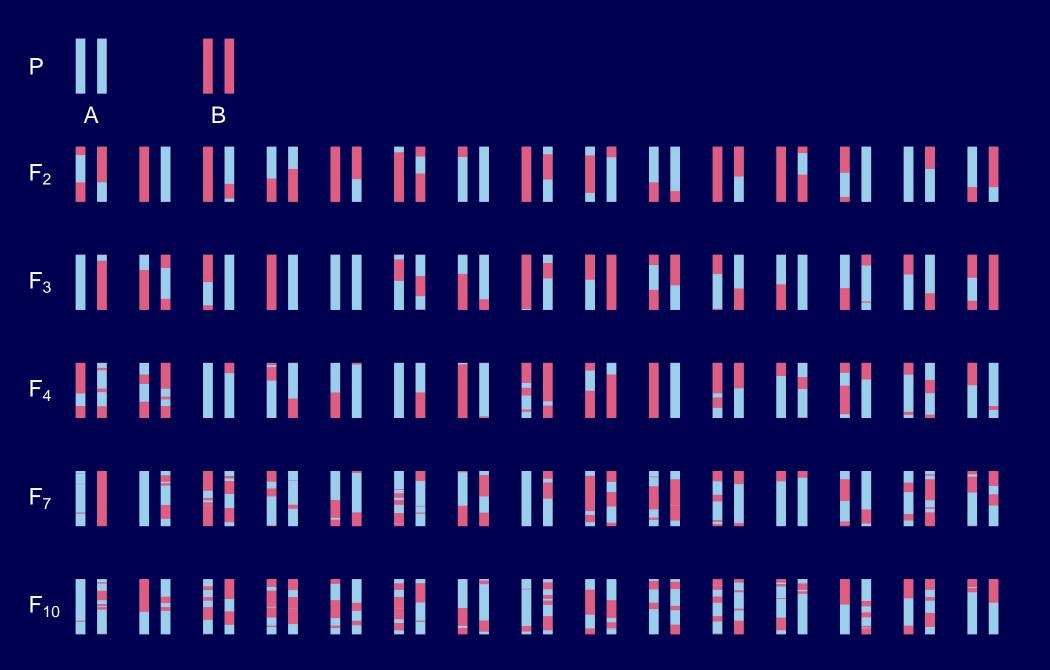
Congenic line / NIL



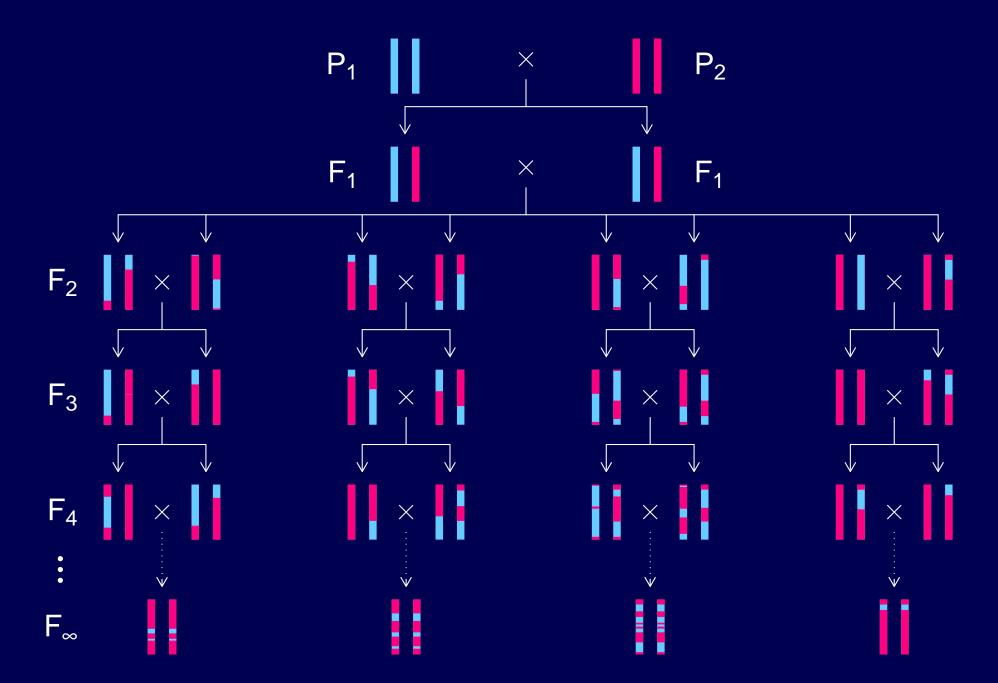
Improving precision

- more recombinations
- more individuals
- more precise phenotypes
- lower-level phenotypes
 - transcripts, proteins, metabolites

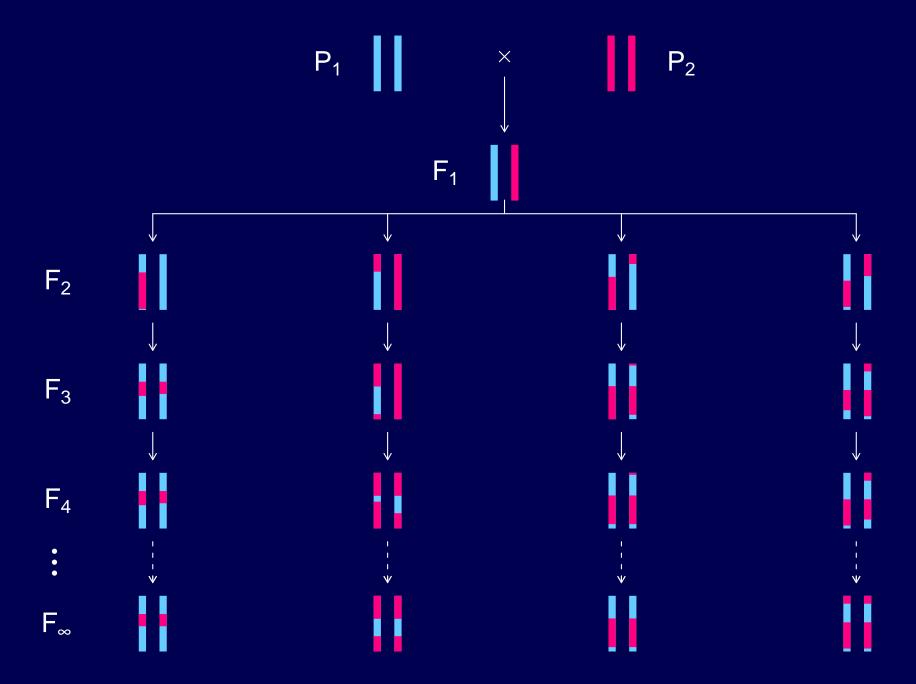
Advanced intercross lines



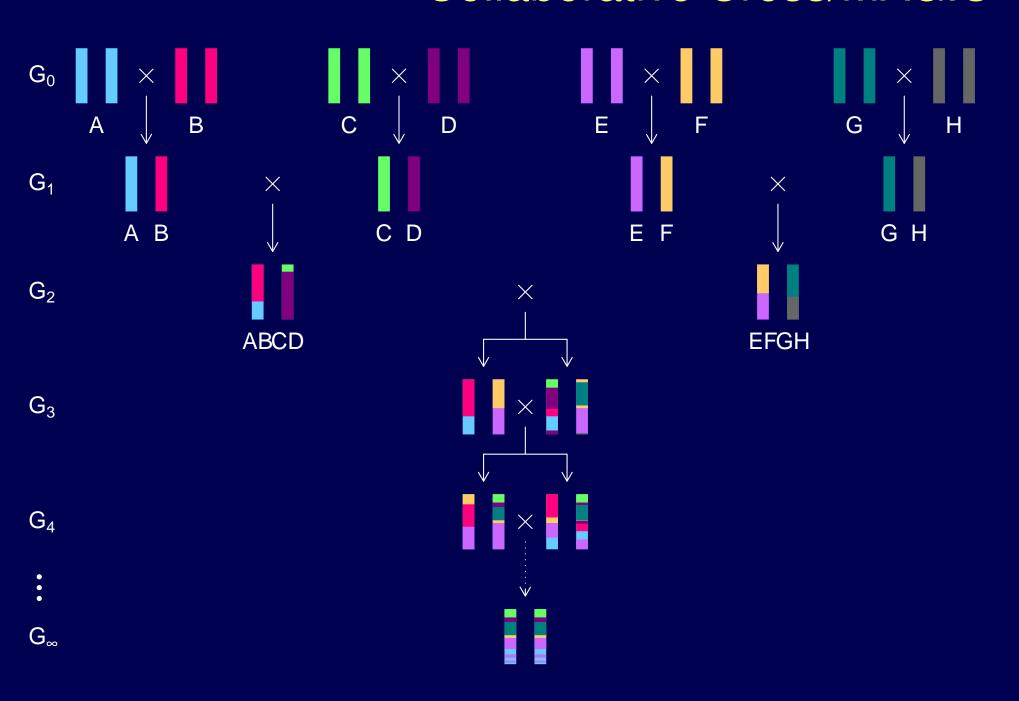
Recombinant inbred lines



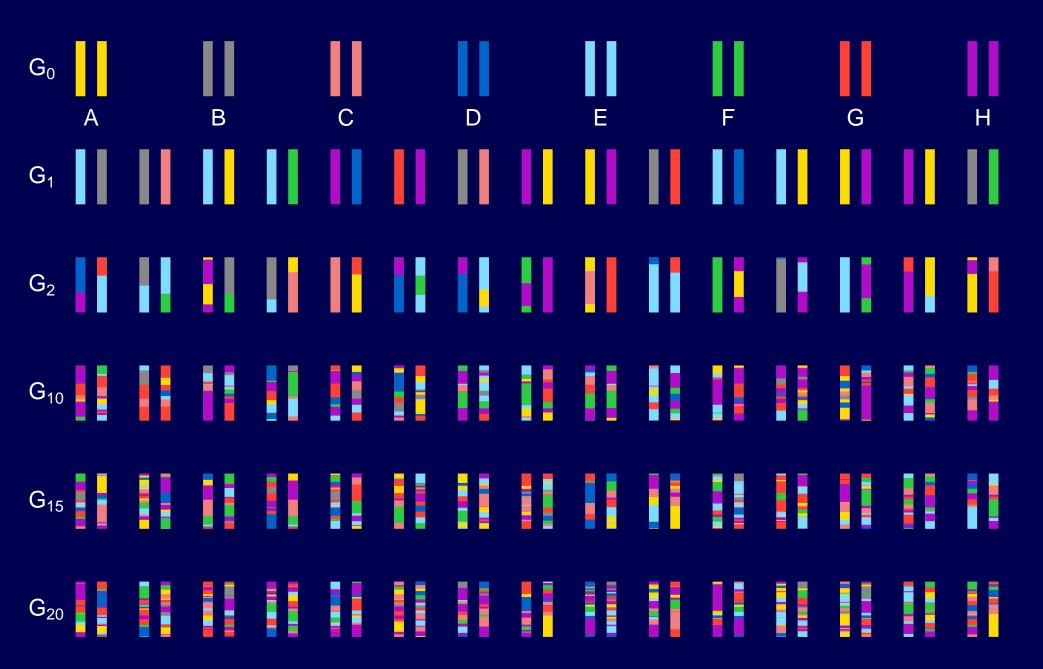
Recombinant inbred lines



Collaborative Cross/MAGIC



Heterogeneous stock



Why R/qtl2?

- High-dimensional data
 genotypes and phenotypes
- More diverse crosses
 especially multi-parent populations
- Linear mixed models
 especially in DO/HS/AIL

$R/qtl \to R/qtl2$

- See kbroman.org/qt12/assets/vignettes/rqtl_diff.html
- New data file formats
- New data structures
- New function names

```
	ext{read.cross()} 
ightarrow 	ext{read\_cross2()} \ 	ext{calc.genoprob()} 
ightarrow 	ext{calc\_genoprob()} \ 	ext{scanone()} 
ightarrow 	ext{scan1()} \ 	ext{}
```

- Different treatment of intermediate calculations
- Use of individual IDs for aligning data
- Order of args when subsetting cross objects

```
cross[chr,ind] \rightarrow cross2[ind,chr]
```

$ightarrow \mathsf{R}$

- convert2cross2()
- summary(), n_ind(), n_mar(), ...
- insert_pseudomarkers()
- calc_genoprob()
- scan1()
- find_peaks()

Linear mixed models

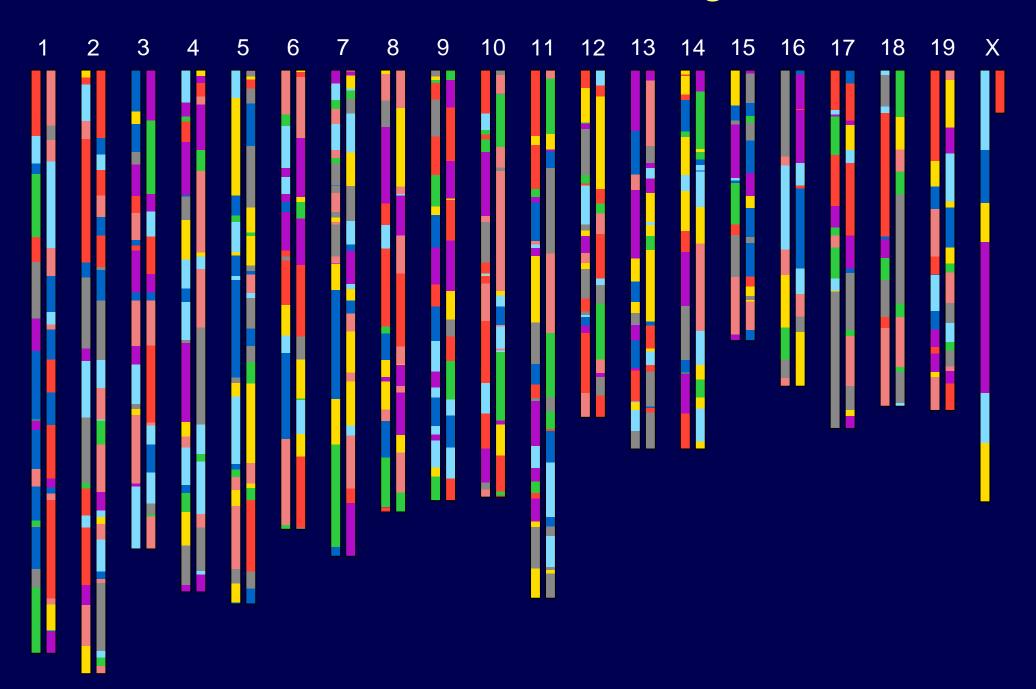
$$y_i = \mu + \sum_k \beta_k q_{ik} + \epsilon_i$$
 $\epsilon_i \sim \mathbf{N}(0, \sigma_e^2)$
 $= \mu + \eta_i + \epsilon_i$ $\eta_i \sim \mathbf{N}(0, \sigma_p^2)$

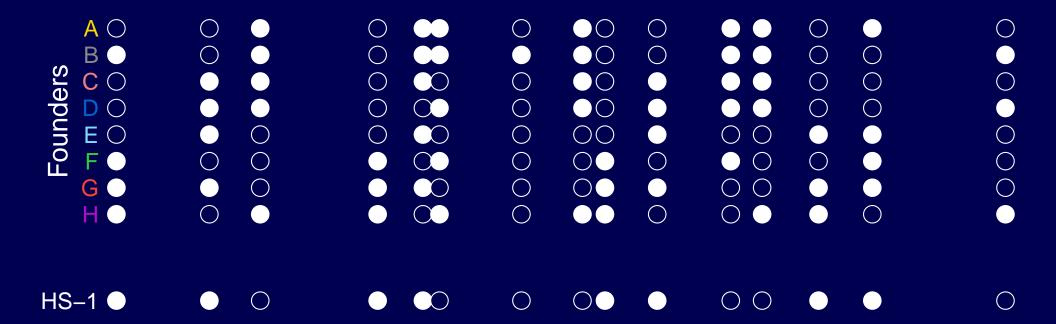
$$\mathbf{COV}(\eta_i, \eta_j) = \sigma_p^2 (2k_{ij})$$

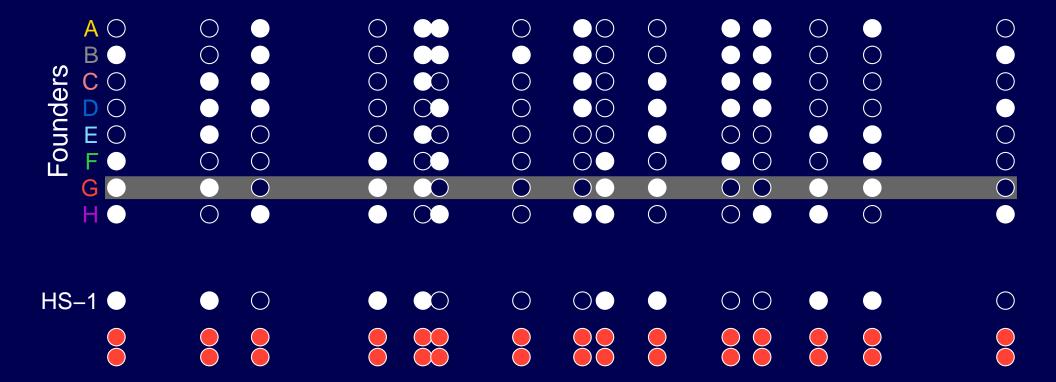
$ightarrow \mathsf{R}$

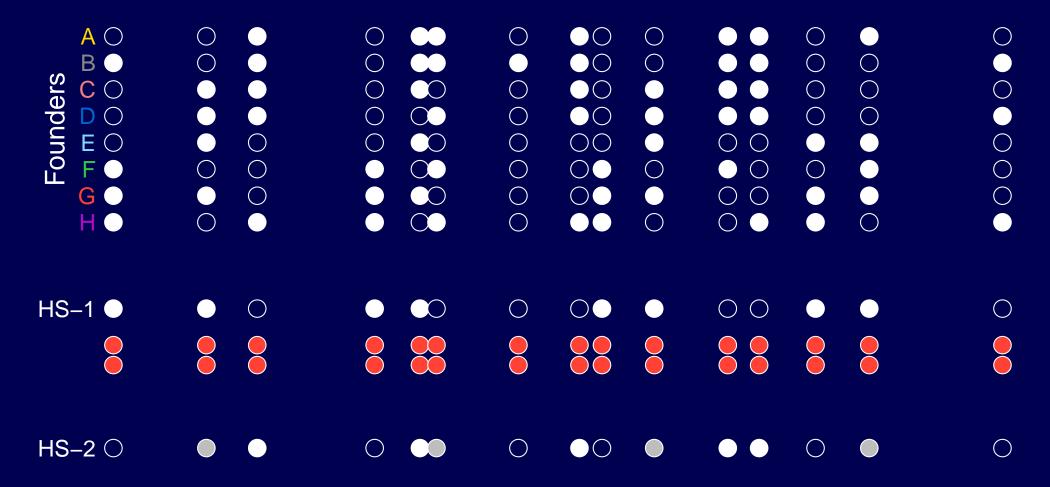
- calc_kinship()
- scan1()

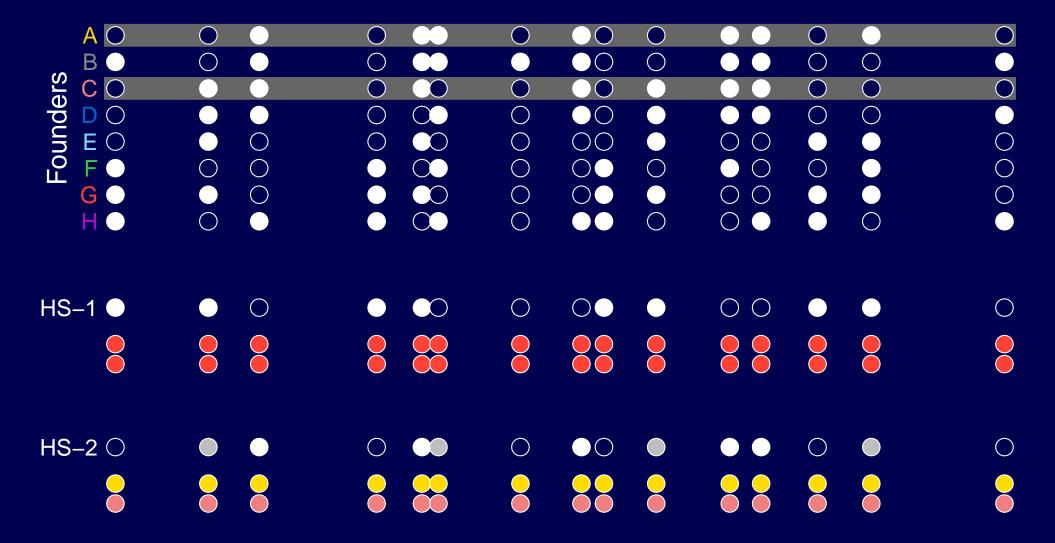
HS genome

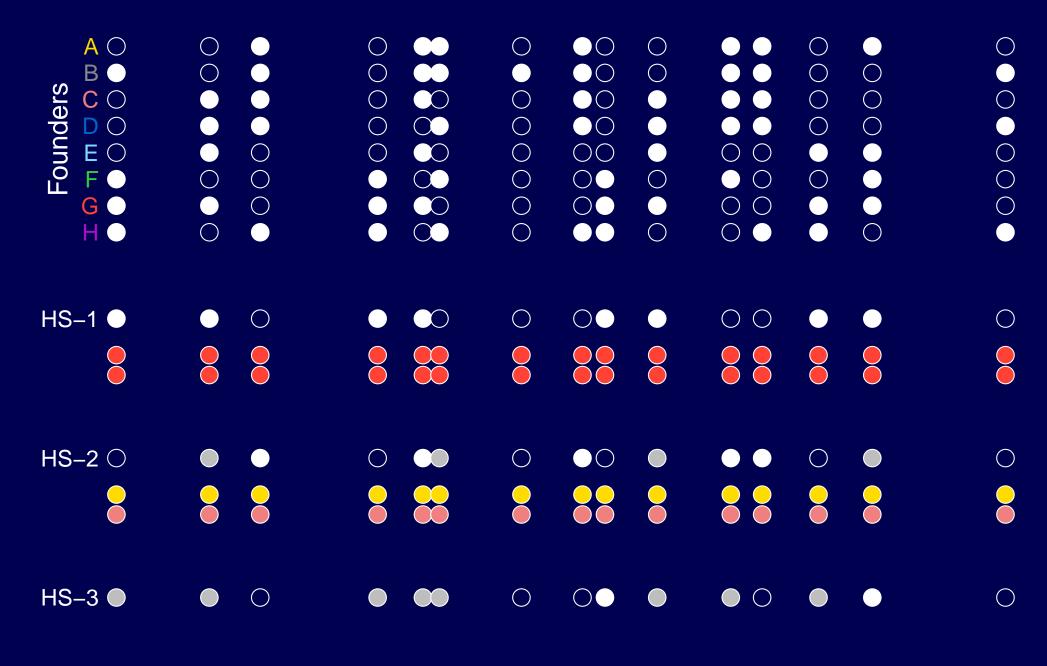


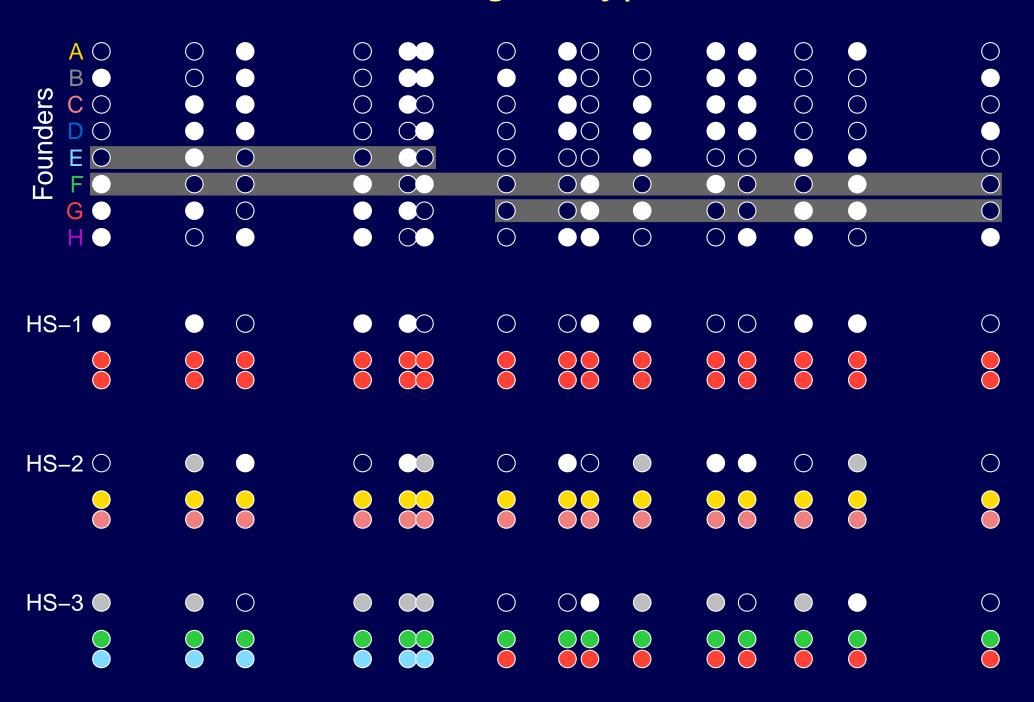






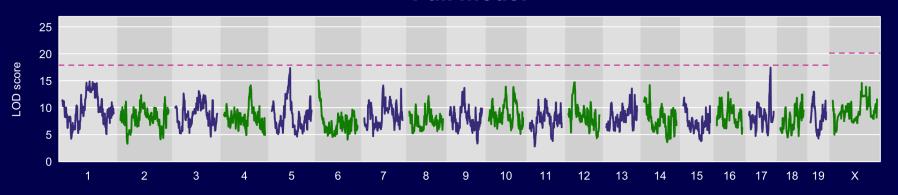




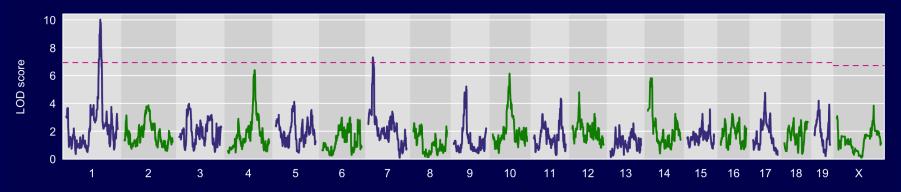


Genome scans

Full model



Additive alleles



SNP associations

